

# Optimizing Zomato Delivery Time through Time Prediction Machine Learning



# Self Introduction



## Education

### Universitas Islam Indonesia (2021-2025) Bachelor in Chemical Engineering

- Designed Preliminary Design of Polyethylene Terephthalate from Ethylene Glycol and Terephthalic Acid



## Work Experience

### Laboratory Assistant Universitas Islam Indonesia

- Delivered 8 modules of practical sessions on Basic Chemical Engineering Laboratory

### Process Engineer Internship PT. Pertamina VI Balongan

- Evaluating splitter column of propylene recovery unit based on real data

**Mita Dian Cahyani**

# Project Overview

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## A/B Testing for E-commerce: Optimizing Speaker Sales Funnels

- **Situation:** A gap in website landing page that result in different revenue
- **Goal:** Efficiently tailor website to drive revenue
- **Action:** Apply website according to customer's preferences
- **Impact:** Increase revenue in 54.4% and conversion rate by 31.5%

## Looker e-Commerce Revenue Growth

- **Situation:** Website traffic is declining
- **Goal:** Identify causes of traffic drop
- **Action:** Optimize AdWords on Firefox in key regions (KY, Pais Vasco, WA, Scotland)
- **Impact:** Targeted traffic growth with efficient ad spend

## Airline Customer Segmentation (Churn Reduction)

- **Situation:** 46.9% of customers are at risk of churn
- **Goal:** Reduce churn through targeted segmentation
- **Action:** K-means clustering to analyze customer behavior
- **Impact:** 10% recovery of at-risk customers saves \$2M

# Table of Contents

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**1) Project Strategy & Objectives**

**5) Modelling**

**2) Business Problem**

**6) App Deployment**

**3) Data Understanding & Preprocessing**

**7) Data Analysis & Insights**

**4) Exploratory Data Analysis**

**8) Recommendation & Impact**

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# Project Strategy & Objectives



## Customer Segmentation

Fast and reliable delivery is a key driver of customer satisfaction in food delivery services. Delays reduce trust, increase complaints, and lower the likelihood of repeat orders.

### Overview

Zomato operates across metropolitan, urban, and semi-urban areas with varying traffic, distance, and weather conditions.

### Goal

To optimize delivery time by using machine learning to predict delivery duration and design strategies that minimize delays.

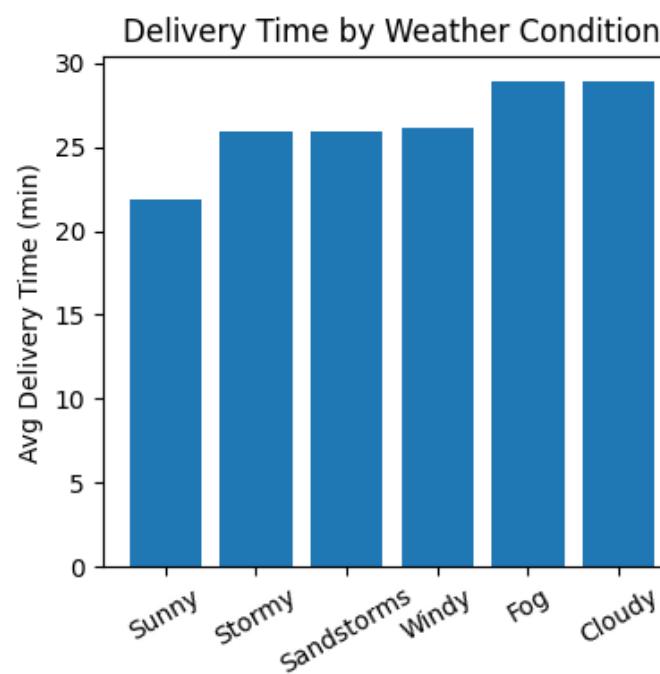
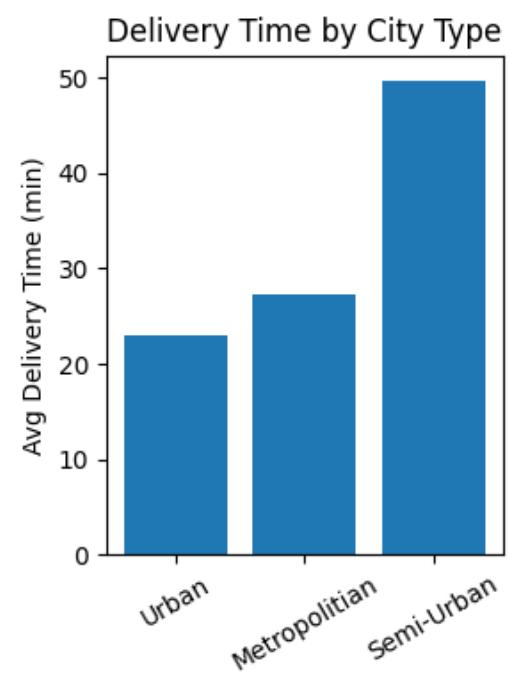
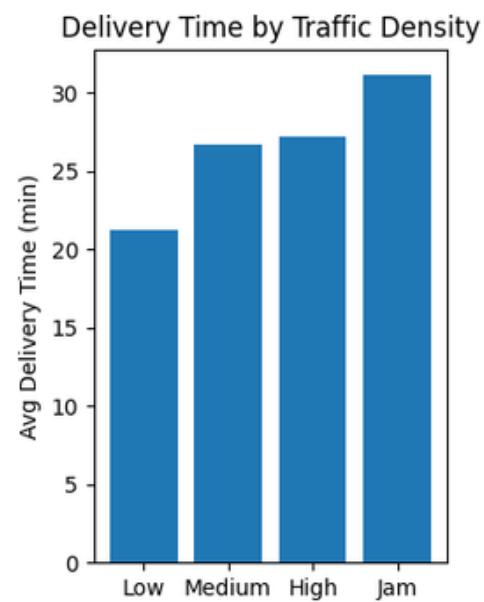
### Objectives

- 1) Delivery Time Prediction:** Develop a machine learning model to predict delivery time in real-time
- 2) Delivery Time Analysis:** Analyse key factors that impact delivery duration.
- 3) Actionable Insights:** Create data-driven operational strategies to reduce delivery time.

# Business Problem



## Problem Statement



**Delivery time is affected by external factors** which may result inconsistent customer experience. This accurate **delivery time prediction is needed to manage orders**

## Why It Matters?



**Customer Satisfaction:** Improves customer satisfaction and retention by setting accurate delivery expectations



**Efficiency:** Reduces operational inefficiencies and late deliveries

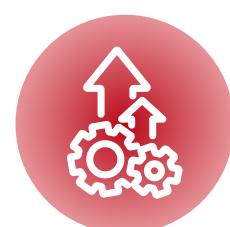
## Stakeholders Benefits



**Customers**  
More accurate delivery time estimates and improved service reliability



**Delivery Partners**  
Better workload distribution and optimized routes



**Business Operations**  
Lower churn, improved efficiency, and higher order completion rates

# Data Understanding & Preprocessing



This dataset details delivery operations by tracking driver profiles, environmental factors like weather and traffic, and geospatial coordinates. It consists of 45,584 rows and 20 columns.

Zomato Delivery Operations Analytics Dataset  
Exploring Delivery Dynamics and Customer Experience on Zomato



Analysis link: [Google Colab](#)  
Tools: [Google colab](#)

## Key Features

Category	Column Names
Temporal	Order_Date, Time_Ordered, Time_Order_picked, Festival
Geospatial	Restaurant & Delivery Lat/Long, City
Environmental	Weather_conditions, Road_traffic_density
Operational	Vehicle_condition, Multiple_deliveries, Type_of_vehicle

## Data Preprocessing

### Missing Value

- Numeric column → median
- Categorical column → mode

### Change Datatype

- Change 'order\_datetime' and 'picked\_datetime' to datetime

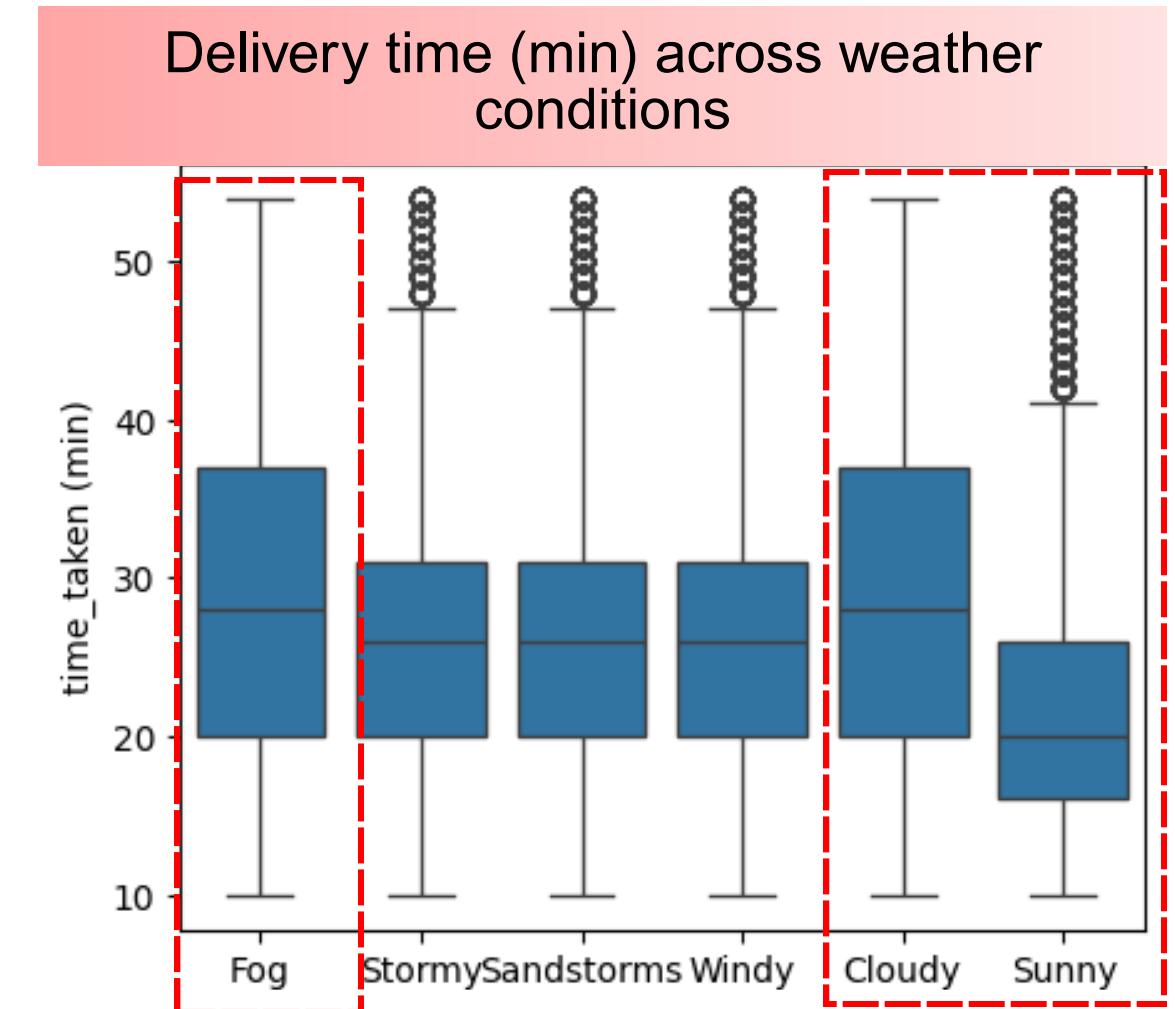
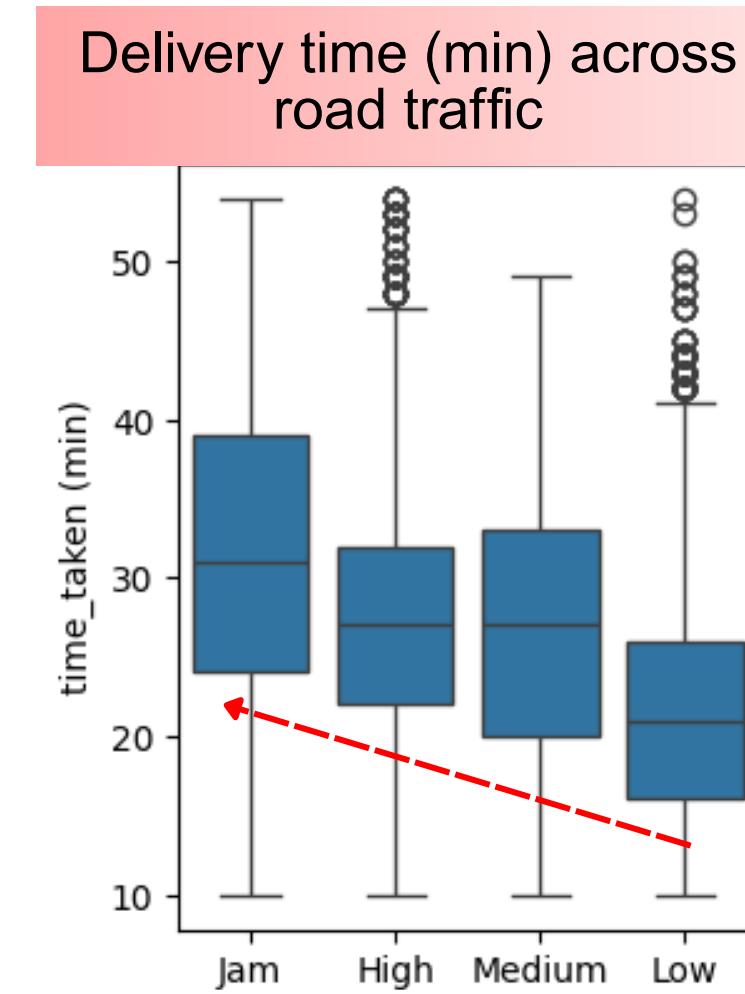
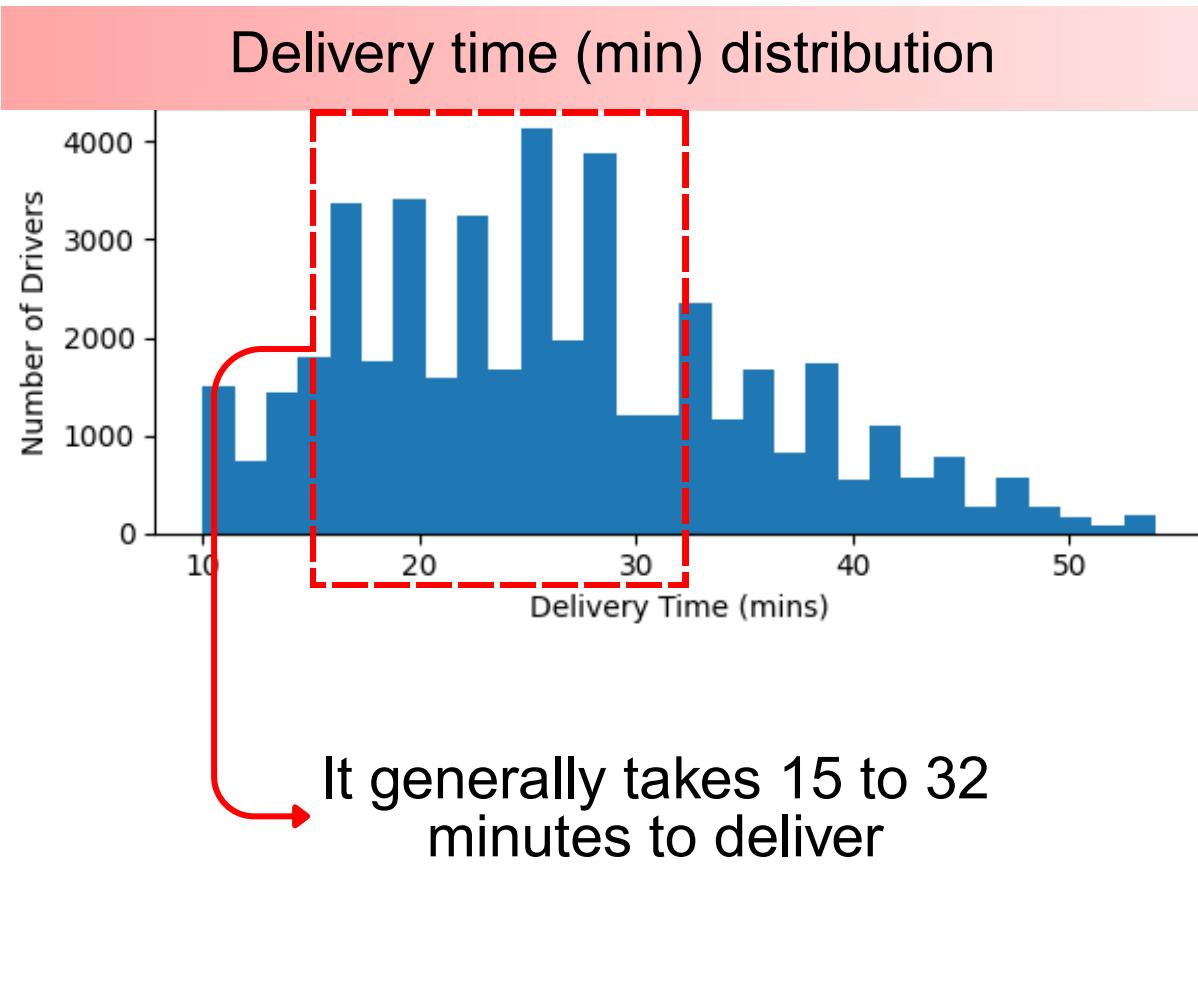
### Feature Engineering

- 'prep\_time\_min' is the time needed for restaurant to prepare order ('picked\_datetime' - 'order\_datetime')
- 'delivery\_dist\_km' is the distance between customer and restaurant (using haversine from restaurant lat/long and delivery lat/long)
- 'order\_hour', 'order\_min' abstracted from 'order\_datetime'

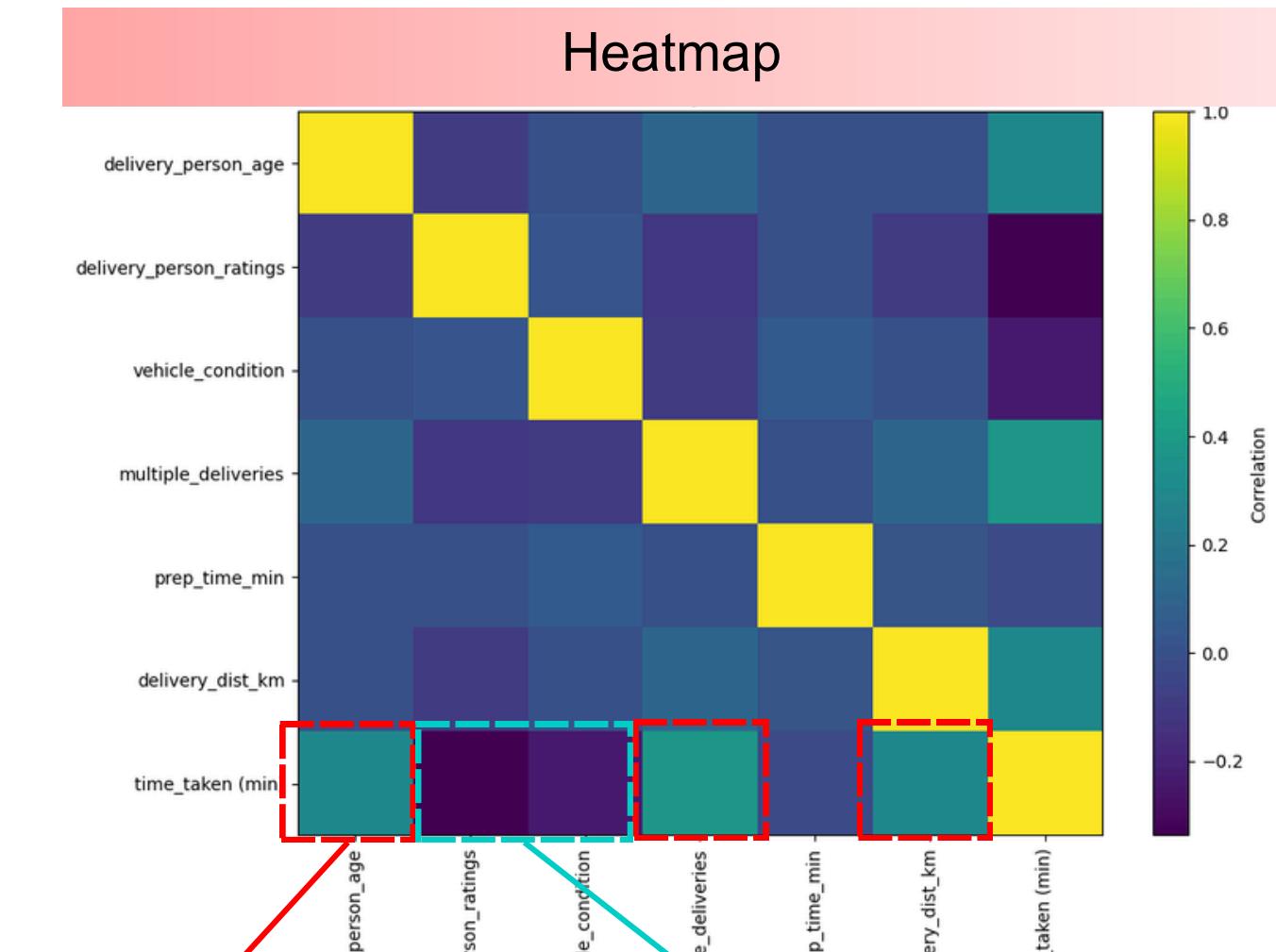
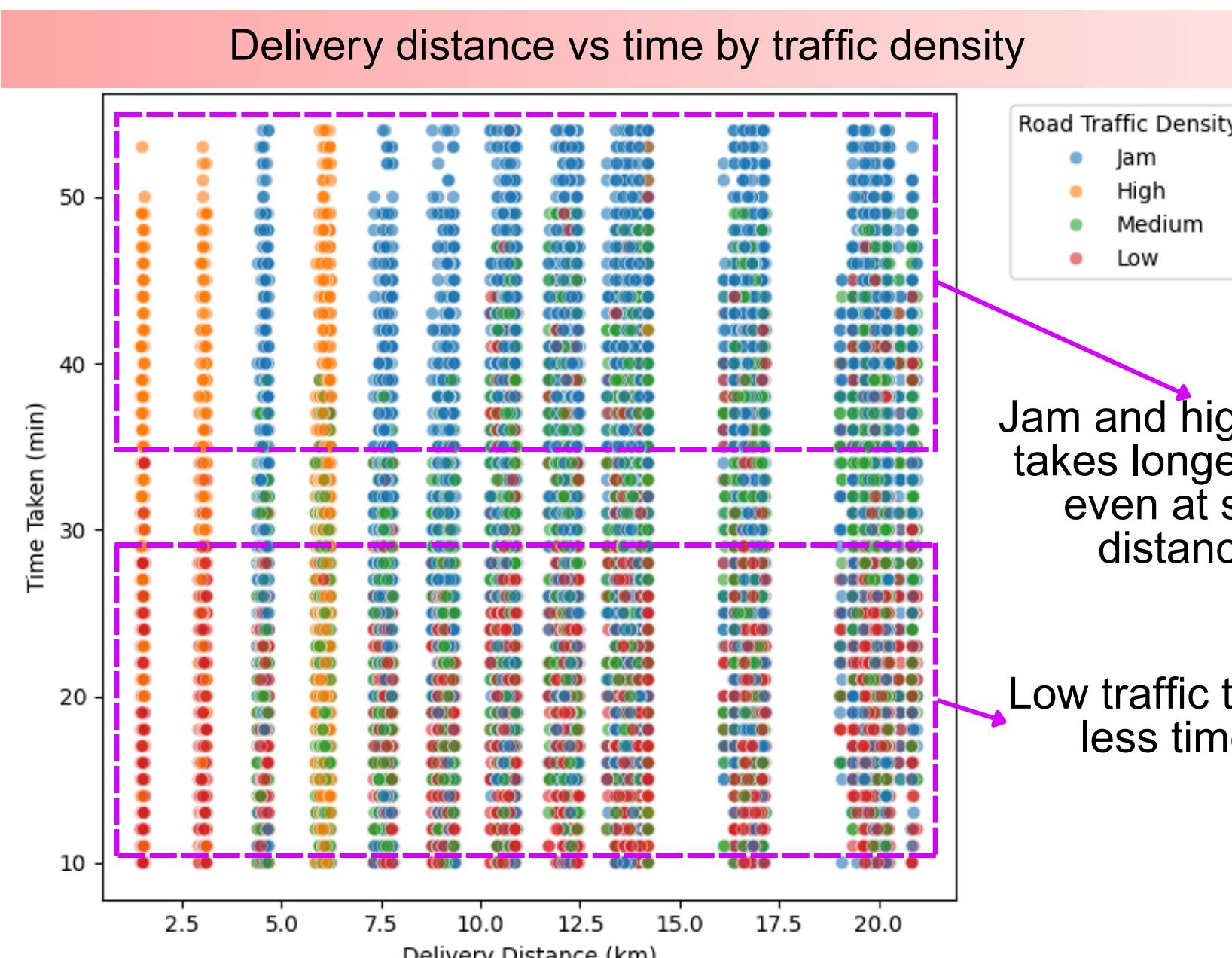
### Missing value percentage

delivery_location_longitude	7.99
delivery_location_latitude	7.99
restaurant_longitude	7.99
restaurant_latitude	7.99
delivery_person_ratings	4.19
delivery_person_age	4.07
time_orderd	3.80
city	2.63
multiple_deliveries	2.18
weather_conditions	1.35
road_traffic_density	1.32
festival	0.50

# Exploratory Data Analysis



# Exploratory Data Analysis



- Delivery time increase when:
- delivery man is older
  - have multiple deliveries
  - farther delivery distance

- Delivery time decrease when:
- Delivery men have lower ratings
  - Vehicle condition is not in best condition

# Modelling



## Model Selection

model	train_MAE	train_RMSE	train_R2	train_MAPE	test_MAE	test_RMSE	test_R2	test_MAPE
RandomForest	2.293537	2.945052	0.901528	10.017312	3.171972	4.028756	0.815543	13.700667
GradientBoosting	3.444700	4.325806	0.787548	15.275560	3.493117	4.376636	0.782313	15.407512
KNN	4.305816	5.463648	0.661083	18.798570	4.751644	6.003080	0.590456	20.739410
Lasso	4.825715	6.070651	0.581594	21.399927	4.823520	6.051273	0.583854	21.344262
Ridge	4.825649	6.070540	0.581609	21.398752	4.823952	6.052136	0.583735	21.344862
LinearRegression	4.825623	6.070539	0.581609	21.398402	4.823969	6.052228	0.583723	21.344622

Why **Random Forest** is chosen?

- Not overfitting
- Good at handling noise
- Low MAE

Main metrics: MAE

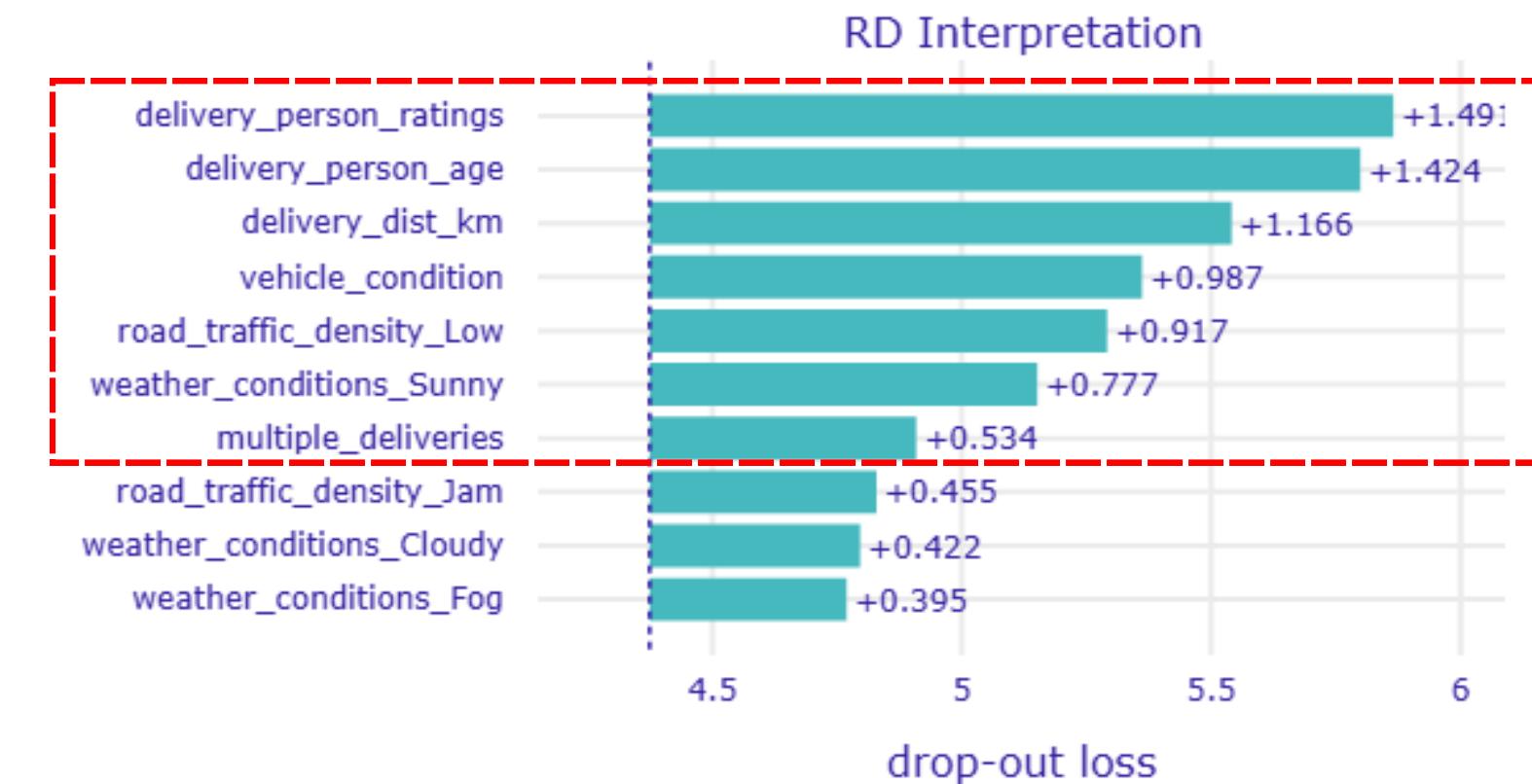
## Hyperparameter Tuning

### Random Forest Parameter

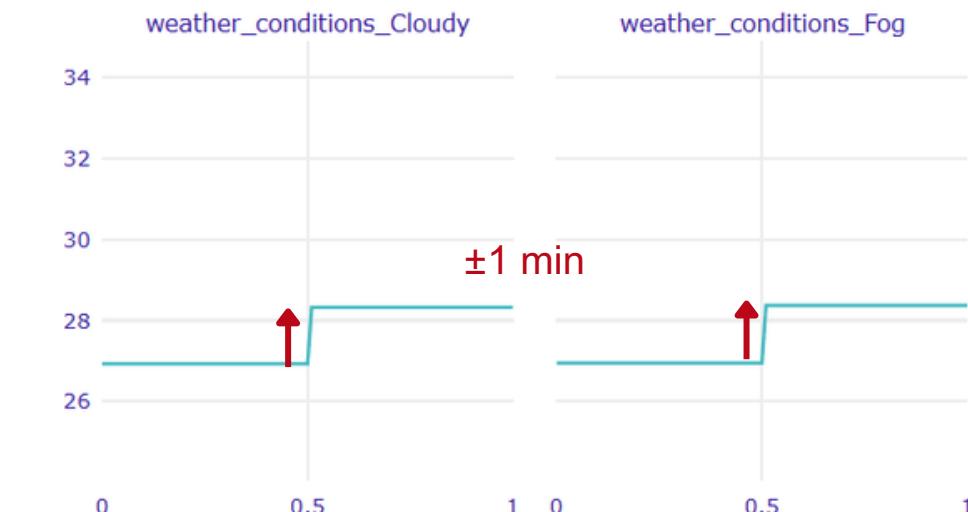
- `max_depth=15`
- `max_features='sqrt'`
- `min_samples_leaf=30`
- `n_estimators=150`
- `n_jobs=-1`
- `random_state=42`

## Variable Importance

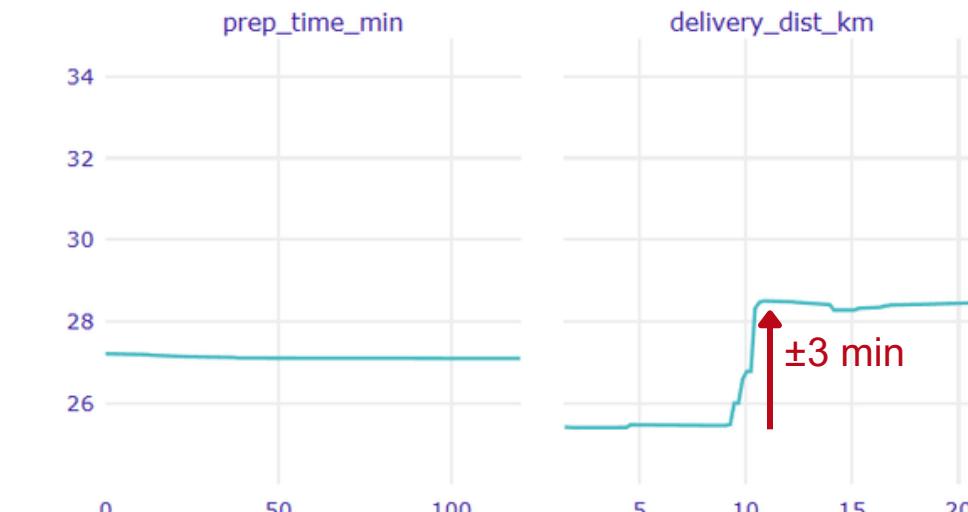
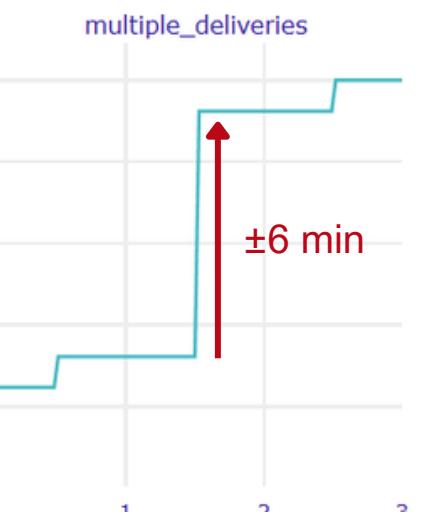
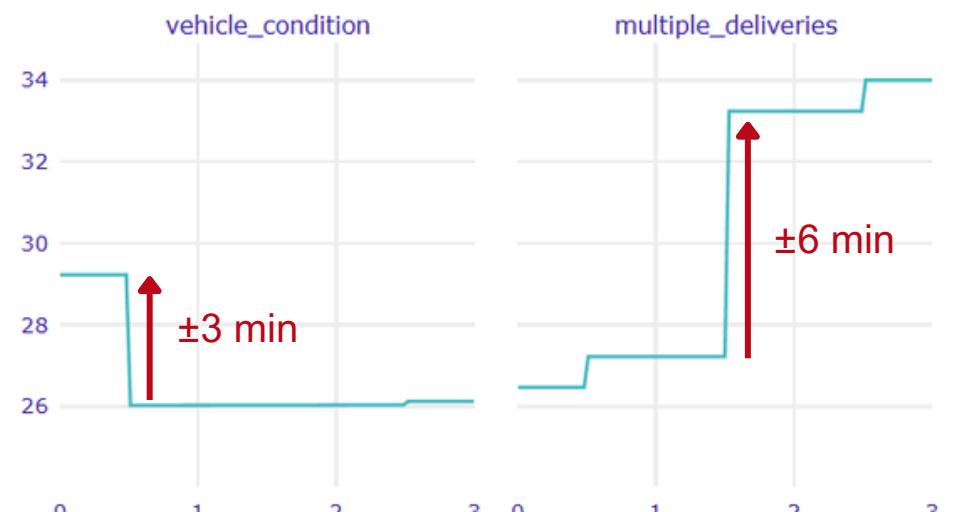
### Interpretation with dalex



# Modelling



Cloudy and foggy weather  $\pm 1$  minute longer on average



# App Deployment



The screenshot shows a Streamlit application interface titled "Enter Delivery Details". The interface includes the following input fields:

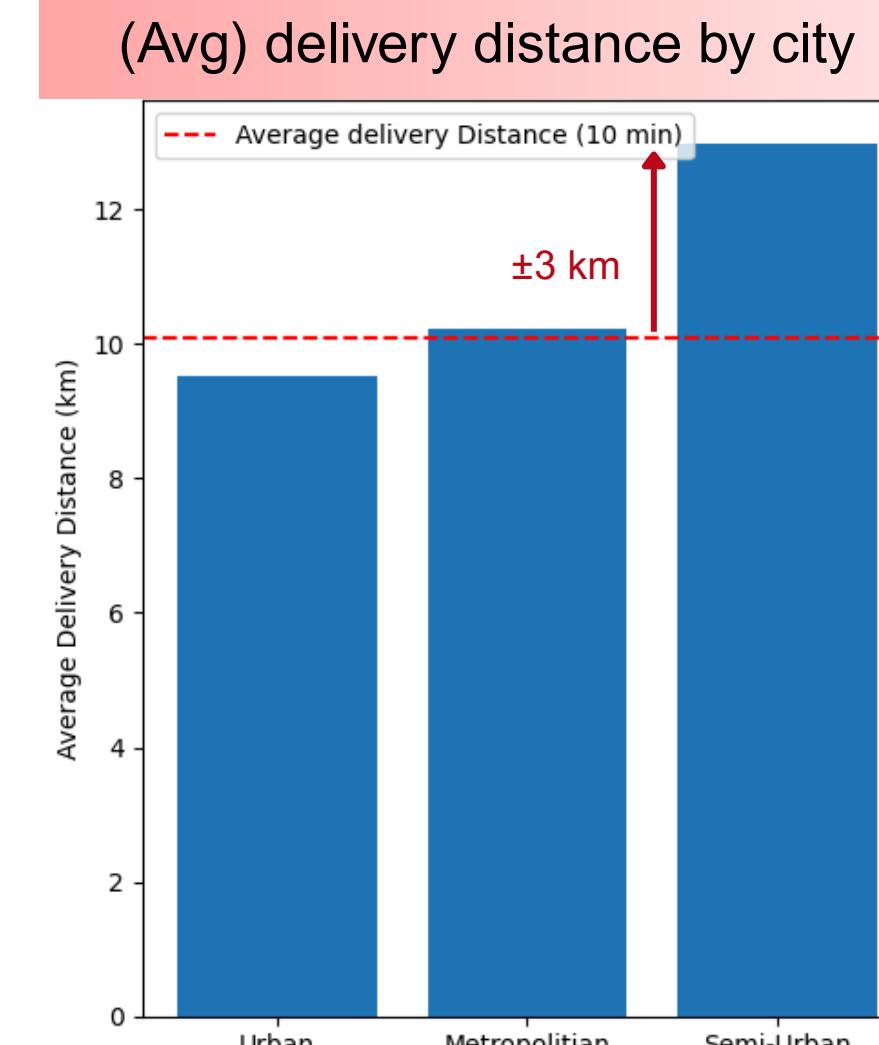
- Distance (km): A slider set to 12,00.
- City Type: A dropdown menu set to "Semi-Urban".
- Delivery Person Age: A slider set to 32.
- Traffic Level: A dropdown menu set to "Jam".
- Vehicle Condition (0 = Worst, 5 = Best): A slider set to 0.
- Weather Conditions: A dropdown menu set to "Fog".

At the bottom right of the form is a "Manage app" button.

Link:

- <https://zomatodeliveryprediction-ycaxu2ay76gam83efmn3yv.streamlit.app/>
- [https://github.com/mitadian/zomato\\_delivery\\_prediction](https://github.com/mitadian/zomato_delivery_prediction)

# Data Analysis & Insights



### Traffic per city (% by row)

road_traffic_density	High	Jam	Low	Medium
city				
Metropolitan	9.859276	32.340827	33.396259	24.403638
Semi-Urban	10.493827	82.716049	0.617284	6.172840
Urban	9.312992	25.974673	41.360056	23.352278

Semi-urban is jammed 82% of time

### Weather per city (% by row)

weather_conditions	Cloudy	Fog	Sandstorms	Stormy	Sunny	Windy
city						
Metropolitan	16.881185	18.319890	16.455008	16.660946	15.153595	16.529375
Semi-Urban	28.395062	20.370370	12.345679	12.345679	12.345679	14.197531
Urban	15.295643	16.272809	16.631768	16.920929	19.124539	15.754312

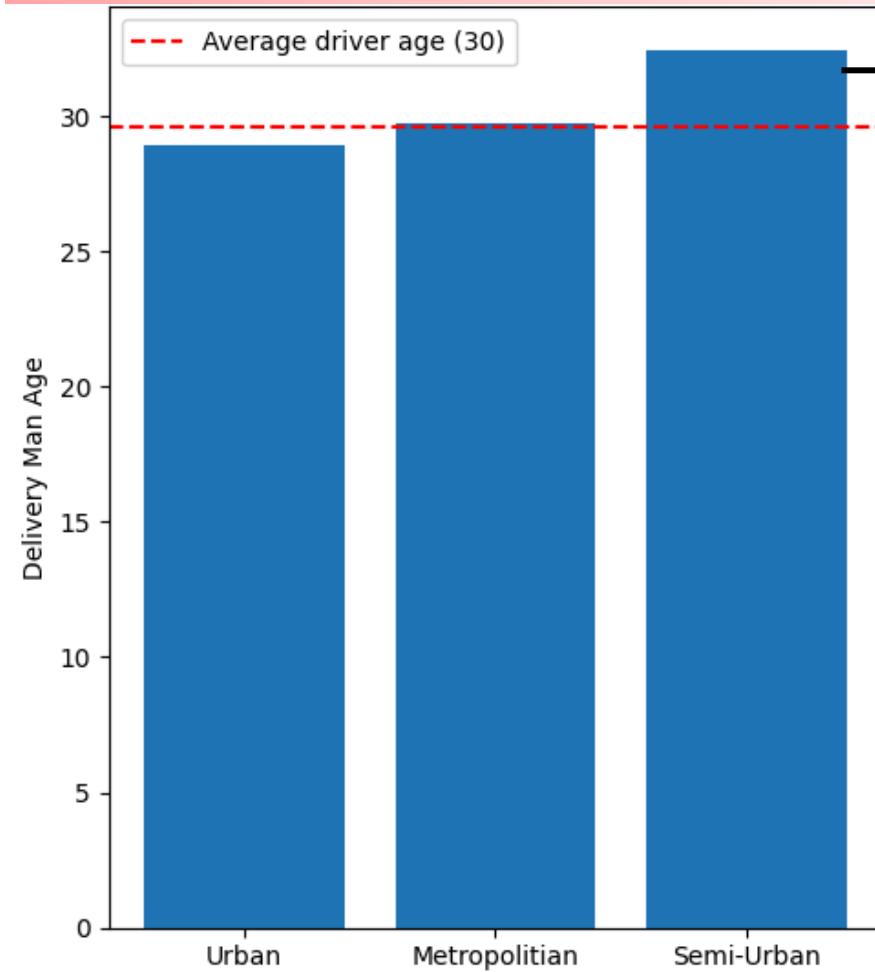
Semi-urban is mostly cloudy and foggy

Semi-urban city is the slowest in delivery which is due to the external factors of far **distance** between restaurant and customer, the **traffic and weather** of the city

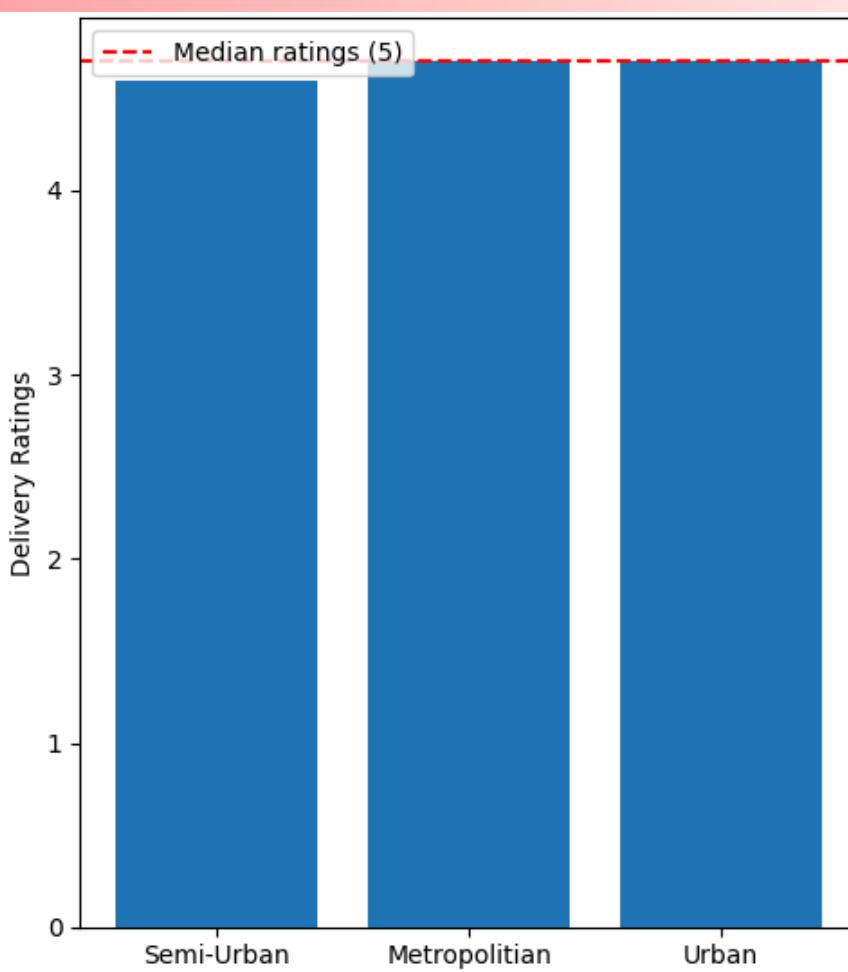
# Data Analysis & Insights



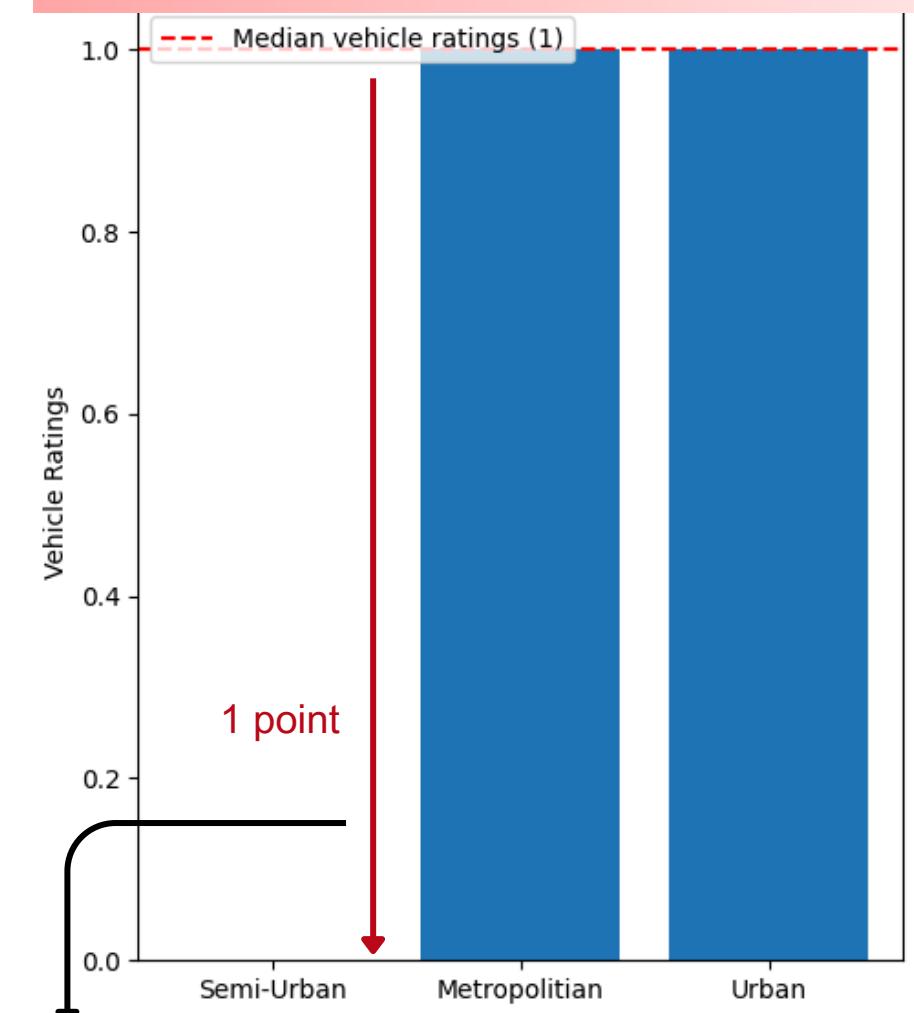
(Avg) driver age by city



(Median) driver ratings by city



(Median) vehicle ratings by city



Semi-urban city is the slowest in delivery which is due to the internal factors, that is generally **older drivers and bad vehicle conditions**

# Recommendations and Impact



## Situation

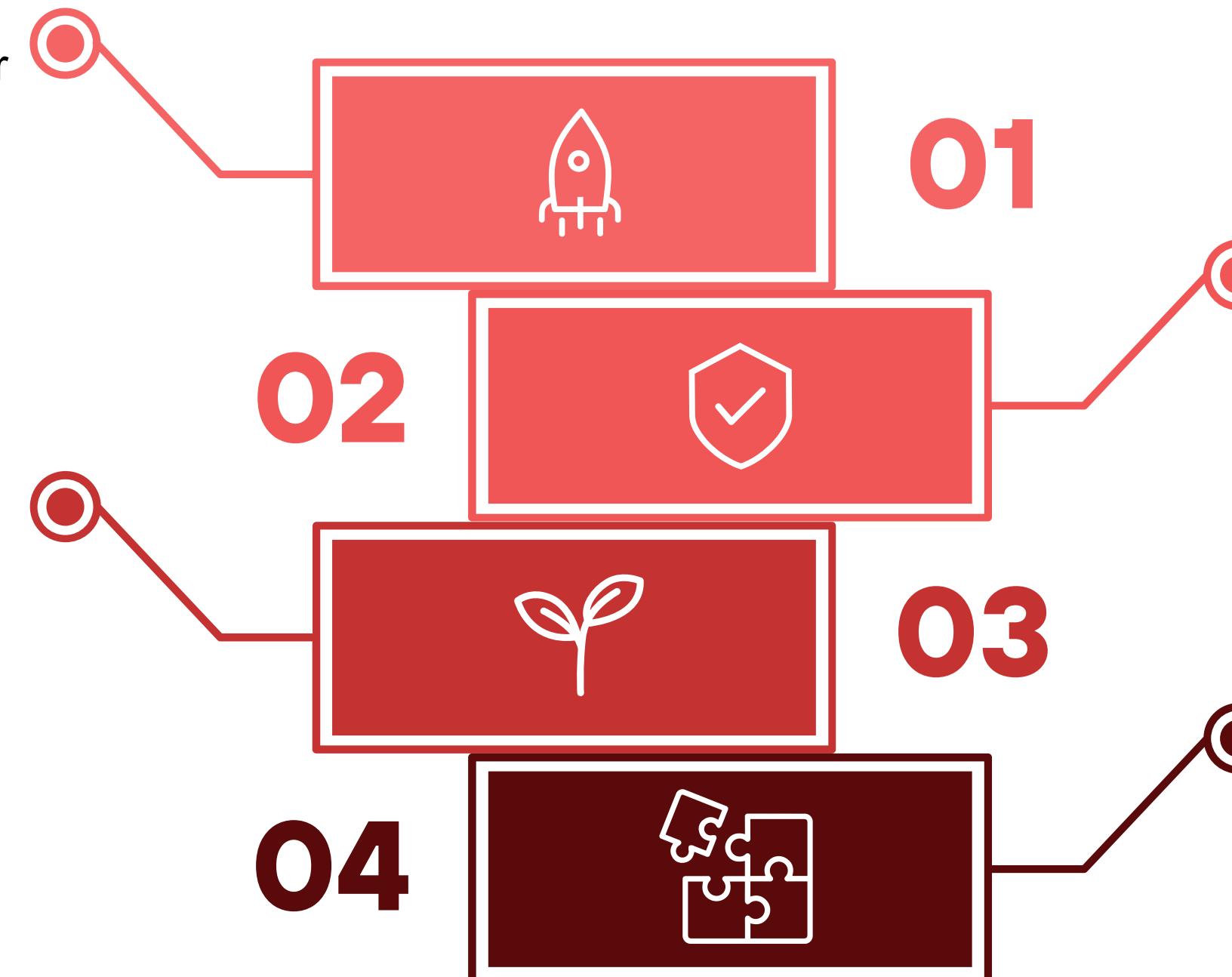
- Semi Urban city are slower in delivery due to delivery distance, weather, traffic, age and vehicle condition

## Recommendations

- Allocate productive drivers in Semi Urban city to handle difficult deliveries
- Apply real time delivery prediction in app to maintain customer trust

note:

- Productive drivers have minimum ratings of 5, maximum age of 30 years and minimum vehicle condition of 1
- Difficult deliveries is delivery distance minimum 10 km, jammed road, cloudy and foggy weather

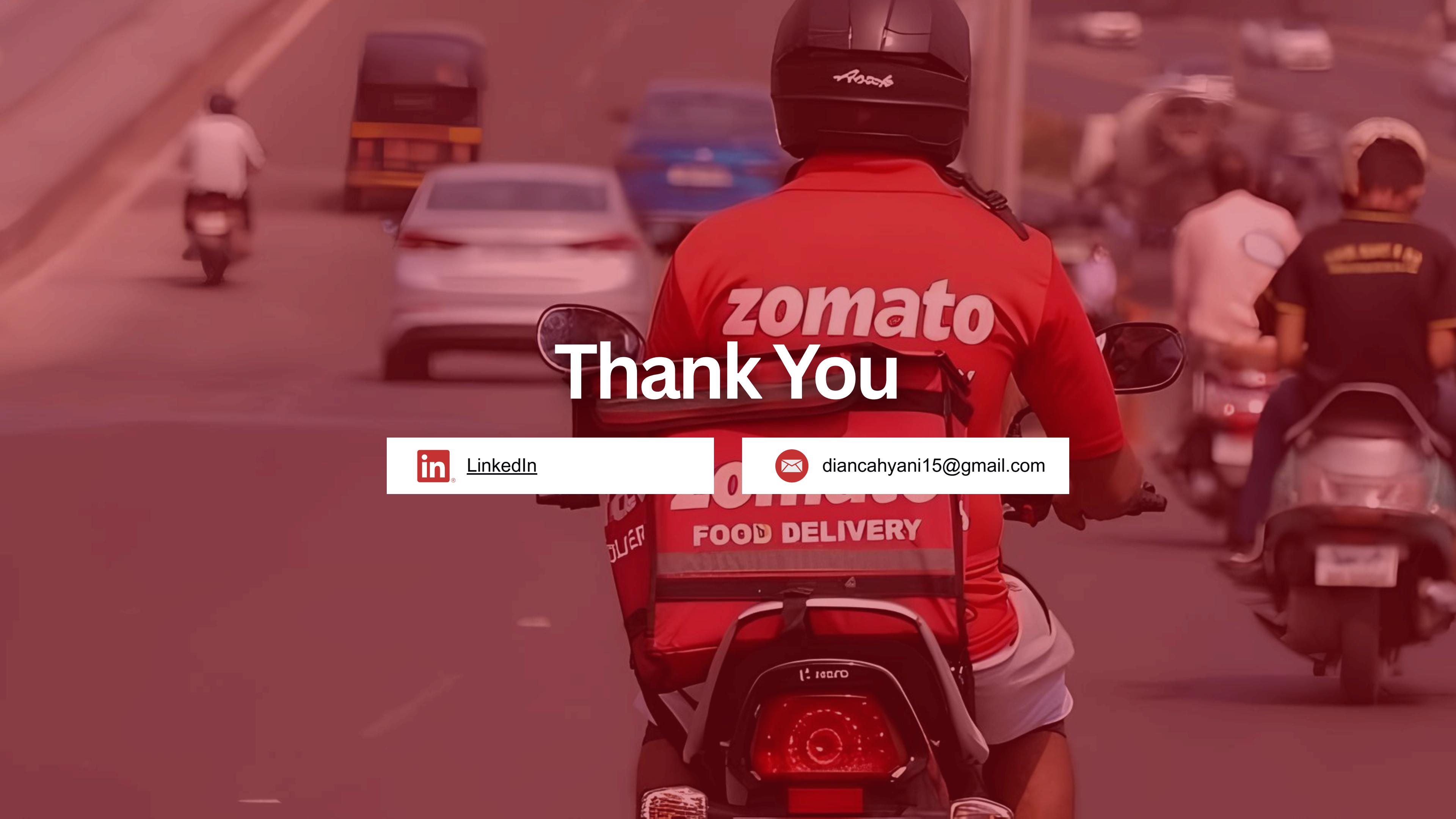


## Problem

- Slow down delivery time affecting customer satisfaction

## Impact

- Reallocating productive drivers can deliver 3 minutes faster for difficult delivery
- Faster delivery times significantly increase customer satisfaction and loyalty (Natasya Rafmadini et al., 2025)



A background photograph shows a Zomato food delivery person wearing a red uniform and helmet, riding a scooter through a busy city street filled with other vehicles and people.

# zomato

# Thank You



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